

Advanced OpenMP Topics

NAS Webinar

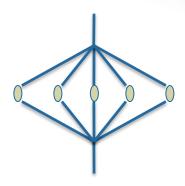
September 12, 2012

NASA Advanced Supercomputing Division

Outline



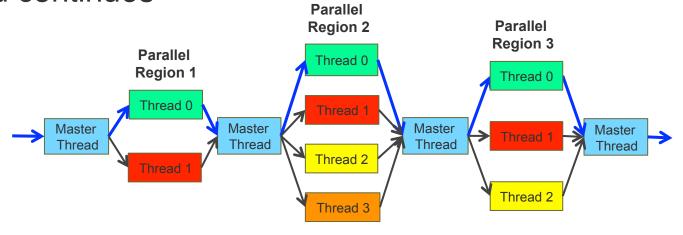
- The *fork-join* model (a refresh)
- Nested parallelism
- OpenMP tasking
 - Task execution model
 - Data scoping
 - Task synchronization
- Performance considerations
- Correctness issues
- Future OpenMP extensions



The Fork-Join Model



- Multiple threads are forked at a parallel construct
 - The *master* thread is part of the new thread team
- Worksharing constructs distribute work in the parallel region
 - for or do, sections, single
- Synchronization primitives synchronize threads
 - barrier, critical, locks
- Threads join at the end of the parallel region and the master thread continues



Nested Parallelism



- Parallel regions can be nested inside another
 - Exploiting parallelism at multiple nesting levels since single level may not be enough

```
#pragma omp parallel for num_threads(2)
for (j=0; j<m; j++) {
    #pragma omp parallel for num_threads(3)
    for (i=0; i<n; i++) {
        c[j][i] = a[j][i] + b[j][i];
    }
}</pre>
Second level parallel region with 3 threads for each outer thread
```

- To enable nested parallel regions
 - OMP_NESTED=true or call omp_set_nested(1)
 - If not, the inner parallel region will be started with a team of one thread
- To set the number of threads
 - Call omp_set_num_threads() or use the num_threads clause
 - OMP_NUM_THREADS=2,3 (OpenMP 3.1)

with a total of 6 threads

Nested Parallelism (cont.)



- Issues with nested parallel regions
 - Performance is a concern
 - Overhead from fork and join
 - Issue with data locality and data reuse
 - Implicit barrier at the end of each inner parallel region
 - Not all compilers (such as PGI compiler) provide the support
- The collapse clause for multiple loops (OpenMP 3.0)
 - Combines closely nested loops into one
 - More efficient than nested parallel regions

```
#pragma omp parallel for collapse(2)
for (j=0; j<m; j++) {
    for (i=0; i<n; i++) {
        c[j][i] = a[j][i] + b[j][i];
    }
}</pre>
```

Tasking in OpenMP



- Limitation of the fork-join model with worksharing constructs
 - Work units statically determined in worksharing constructs
 - No easy method to dynamically generate work units
 - Lack of support for recursive algorithms
 - For example, no easy way to traverse a tree in parallel
- Tasking model
 - Introduced in OpenMP 3.0
 - Complimentary to the thread-centric model
 - Ability to express parallelism for recursive algorithms, pointer chasing, which are commonly encountered in C/C++
 - Constructs for task generation and task synchronization
 - Concept of task switching

Basic Task Concept

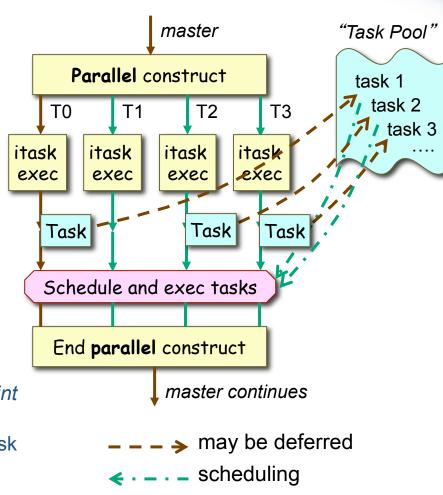


- OpenMP task
 - A code entity including control flow and its data environment, executed by a thread
- Implicit and explicit tasks
 - Implicit tasks generated via the parallel directive
 - Explicit tasks generated via the task directive
- Task synchronization
 - The **taskwait** directive to wait for all *child tasks* of the current task
 - Implicit or explicit barriers to wait for all explicit tasks
- Data environment is associated with tasks except for threadprivate storage
- Locks are owned by tasks
 - Set by a task, unset by the same task

Task Execution Model



- Starts with the master thread
- Encounters a parallel construct
 - Creates a team of threads, id 0 for the master thread
 - Generates implicit tasks, one per thread
 - Threads in the team executes implicit tasks
- Encounters a worksharing construct
 - Distributes work among threads (or implicit tasks)
- Encounters a task construct
 - Generates an explicit task
 - Execution of the task could be deferred
- Execution of explicit tasks
 - Threads execute tasks at a task scheduling point (such as task, taskwait, barrier)
 - Thread may switch from one task to another task
- At the end of a parallel construct
 - All tasks complete their execution
 - Only the master thread continues afterwards



- implicit tasks cannot be deferred
- explicit tasks could be deferred

Thread versus Task



- Threading model
 - Thread and work (or task) go together
 - No concept of deferred execution
- Tasking model
 - Task generation and task execution are separate
 - There is no direct control on when a task gets executed
 - Thread is an execution engine
 - It is a more dynamic environment
- OpenMP supports both models

Tasking Example: Fibonacci Number



```
int fib(int n)
                                   Explicit tasks with proper
                                   data sharing attributes
 int x, y;
  if (n < 2) return n;
  #pragma omp task shared(x)
   x = fib(n-1);
  #pragma omp task shared(y)
   y = fib(n-2);
  #pragma omp taskwait
 return(x+y);
                                 Ensure calculations for x
                                 and y are done and storage
                                 does not disappear
    Single thread generates
    tasks, but multiple
    threads execute tasks
```

The code builds a binary task tree. Parallelism comes from the execution of tasks on the leaf nodes.

But don't expect any performance from this version!

Data Sharing in Tasks



"p" is private and "n" is shared in

- Default sharing attribute rules
 - Shared for implicit tasks
 - For explicit tasks
 - If a variable is determined to be shared in the parellel region, default is shared
 - Otherwise, default is firstprivate (to avoid out-of-scope data access)
- Use data sharing clauses explicitly, in particular if you are not sure
 - **shared**, **private**, **firstprivate**, etc.

```
the parallel region
node_t *node_head, *p;
int n = 40;
#pragma omp parallel private(p)
 #pragma omp master
  p = node_head;
  while (p) {
     #pragma omp task
      process(p,n);
                            "p" is firstprivate
      = p->next;
                           and "n" is shared
                           for the task
 #pragma omp taskwait
```

Example of pointer chasing

Common Problems in Using OpenMP



- Code is not scaling possible issues:
 - Overhead of OpenMP constructs
 - Granularity of work units
 - Remote data access and NUMA effect
 - Load imbalance
 - False sharing of cache
 - Poor resource utilization
- Parallel code gives a slightly different result than the serial code
 - Understanding parallel reduction
- Code crashes or gives different results from run to run
 - Stack size limitation
 - Data race

Overhead and Granularity



- Overhead from OpenMP constructs
 - Fork-join of threads
 - Barrier
 - Creation and scheduling of tasks
 - May be measured with the EPCC microbenchmarks
- Not enough granularity in work unit
- Possible solutions
 - Increase work and exploit parallelism at coarser level
 - Merge parallel regions if possible
 - Avoid barrier if possible (e.g., with *nowait* clause)
 - Use atomic over critical or reduction

Reducing Overhead



Example 1

```
#pragma omp parallel for
  for (i=0, i<n; i++)
    a[i] = b[i] + c[i];

#pragma omp parallel for
  for (i=0; i<n; i++)
    d[i] = e[i] + f[i];</pre>
```



```
#pragma omp parallel
{  #pragma omp for nowait
  for (i=0; i<n; i++)
    a[i] = b[i] + c[i];
  #pragma omp for nowait
  for (i=0; i<n; i++)
    d[i] = e[i] + f[i];
}</pre>
```

Example 2

```
for (i=0; i<m; i++){
    #pragma omp parallel for
    for (j=0; j<n; j++)
        a[i][j] += a[i-1][j]
        + a[i+1][j];
}</pre>
```



```
#pragma omp parallel private(i)
for (i=0; i<m; i++){
    #pragma omp for
    for (j=0; j<n; j++)
        a[i][j] += a[i-1][j]
        + a[i+1][j];
}</pre>
```

- Merge parallel regions
- Use *nowait* if no data dependence between worksharing regions

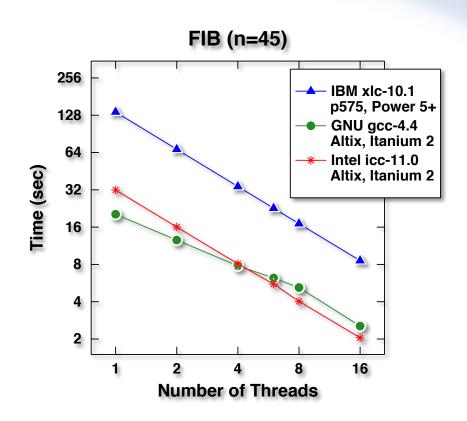
Fibonacci Number – Increased Granularity



```
int fib(int n)
{
   int x, y;
   if (n < 2) return n;
   if (n < 30)
    return (fib(n-1)+fib(n-2));
   #pragma omp task shared(x)
    x = fib(n-1);
   #pragma omp task shared(y)
   y = fib(n-2);
   #pragma omp taskwait
   return(x+y);
}</pre>
```

IF condition added to avoid finegrained tasks and increase task granularity

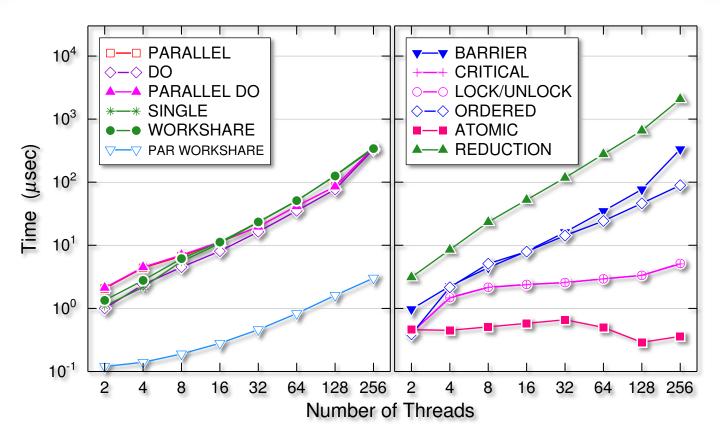
Each task performs some amount of work!



Performance from the naïve version is not shown here – it is more than 10-fold worse and does not scale

EPCC Microbenchmark Results



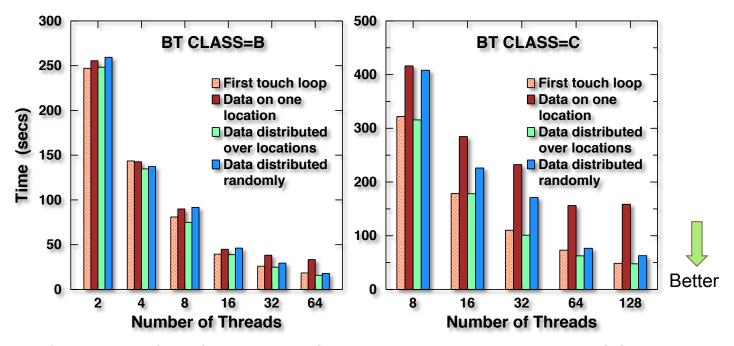


- Measure extra time spent (or overhead) by each OpenMP construct as a function of thread counts on the SGI Altix
- Intel OpenMP compiler was used
- Constructs such as parallel, reduction, barrier have very large overhead

Remote Data Access and NUMA Effect



- Remote data access is more expensive
 - May cause memory access bottleneck
- Possible solutions
 - Use thread-local data (private or threadprivate) if possible
 - Add the first touch loop



- Performance of BT from the NAS Parallel Benchmarks on the SGI Altix
- Four types of data layout based on how data are initially distributed

Other Performance Issues



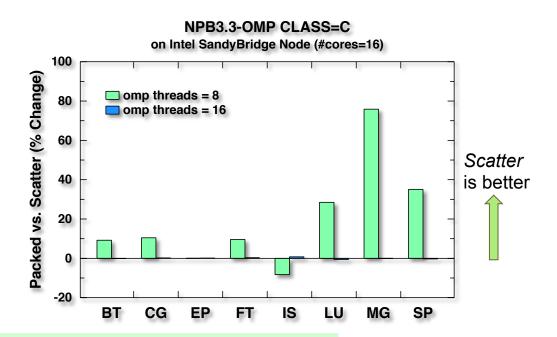
- Load imbalance
 - Try the *dynamic* loop schedule
 - Increase iteration space by using the *collapse* clause for nested loops
- False sharing
 - Caused by multiple threads updating data in the same cache line
 - Work-around
 - Pad array dimension of shared data
 - Use private data if possible
- A good practice
 - Use omp_get_wtime() to get timing profile for code sections in question

Thread-Processor Binding



- Or thread affinity
 - May improve performance by reducing OS scheduling overhead and improving resource utilization
 - Reduce run-to-run timing variation
 - But no standard way currently to control the affinity setting
 - For Intel compiler, set KMP_AFFINITY={scatter,compact..}

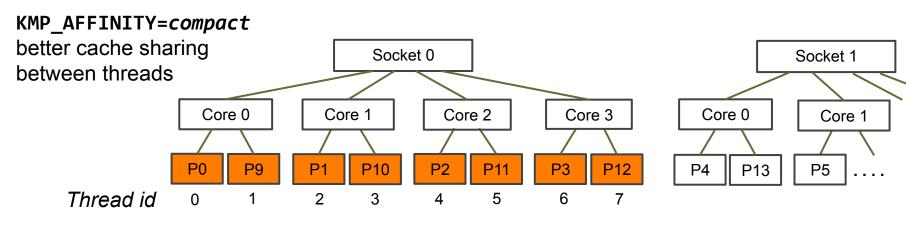
Example of using thread binding from two types of affinity settings to improve resource utilization

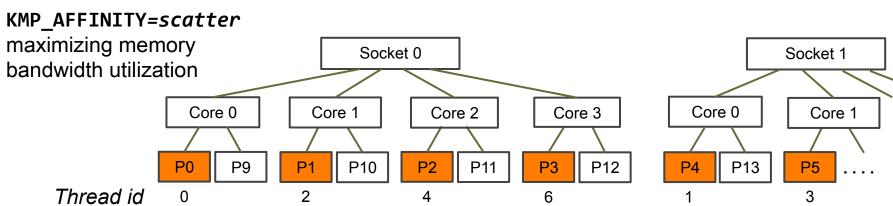


Thread Affinity Types



Examples of Intel Compiler, **OMP_NUM_THREADS=8**, two quad-core sockets





- "scatter" usually gives better results for most cases

KMP_AFFINITY=explicit, proclist=[0-7] user specifies the proc list explicitly

For more details, see www.nas.nasa.gov/hecc/support/kb/60/

Code Correctness Issues



Parallel reduction

- May not be bit reproducible as the serial result
- Mathematically associative: (x + y) + z = x + (y + z), but machine accuracy is limited for floating point
- Use double precision over single precision for reduction variables

Some common programming errors

- Incorrect variable scoping
- Accessing reduced variable without a barrier
- Master versus single
 - Master doesn't have a barrier, but single does
- Race condition

Race Condition



- Commonly encountered in shared memory programming
 - Results are not deterministic
 - Unintentional (programming error), intentional (one thread polling a flag that is updated by another thread)
- Occurs when all the following hold
 - Multiple threads access the same memory location concurrently
 - One of the access is write
 - Access is not protected (e.g., by critical construct)

```
Updating private variable "tid" is OK

#pragma omp parallel private(tid)
{

tid = omp_get_thread_num();

n = tid;
}
```

Updating shared variable "n" from multiple threads causes a race. Race condition should be avoided by all means.

Code Correctness Issues (cont.)



Code crashes

- Caused by programming errors
 - Debugging the code with a debugger (gdb, totalview, etc.)
- Runtime stack size limitation
 - Default thread stack size can be easily exhausted
 - Reset stack size for master threads via shell command

```
limit stacksize unlimit (csh)
ulimit -s unlimited (sh)
```

Reset stack size for worker threads via environment variable

```
setenv OMP_STACKSIZE 12m (csh)
export OMP_STACKSIZE=12m (sh)
```

Software Tools



- Correctness checking
 - Variable scoping
 - "Auto" scoping supported by the Oracle OpenMP compiler
 - Race condition detection
 - Intel Thread Checker (or Parallel Inspector)
 - Oracle Thread Analyzer
- Performance tools
 - Compiler feedback
 - Profiling tools
 - ompP (UCB), PerfSuite (NCSA), Vtune (Intel), TAU (U.Oregon), etc.
- Parallelization assistant
 - Compiler auto-parallelization
 - Semi-automatic parallelization tools (CAPO/Parawise)

Future OpenMP Extensions



- Work in progress within the OpenMP language committee
 - Public draft of the 4.0 specification by the end of the year
- New features under consideration
 - User-defined reduction
 - Error handling
 - The cancel construct for parallel and worksharing
 - Cancellation points
 - Fortran 2003 support
 - Thread affinity
 - Logical processor units via the OMP_PLACES environment variable
 - Affinity policy (compact, scatter, master) for threads in parallel regions
 - Handling thread affinity in nested parallel regions
 - Atomic construct for sequential consistency
 - atomic seq_cst

Support for Accelerator Devices



- Such as GPUs, Intel Xeon Phi (MIC)
 - Many cores, large amount of parallelism
 - Disjoint device memory from the host
- Programming models
 - Low level models (CUDA, OpenCL) exist, but hard to use
 - High level models are being developed
- OpenACC model (for GPUs)
 - Based on the PGI Accelerator programming model, defined by multivendors (www.openacc-standard.org)
 - Using compiler directives, as in OpenMP
 - Offloading work to the device
 - Data transfer between the host and the device
 - Intend to merge into OpenMP eventually

Summary



- OpenMP provides a programming model for shared memory systems
- Compilers with OpenMP support are widely available
- The tasking model opens up opportunities for a wider range of applications
- Several issues to consider for developing efficient OpenMP codes
 - OpenMP overhead
 - Data locality
 - In some cases trade-off between easy of use and performance
 - With some extra effort, scalability can be achieved in many cases

References



- OpenMP specifications
 - www.openmp.org/wp/openmp-specifications/
- Resources
 - www.openmp.org/wp/resources/
 - www.compunity.org/
- Benchmarks
 - OpenMP Microbenchmarks from EPCC (www.epcc.ed.ac.uk/research/openmpbench)
 - NAS Parallel Benchmarks (www.nas.nasa.gov/publications/npb.html)
- Porting applications to Pleiades
 - www.nas.nasa.gov/hecc/support/kb/52/
 - www.nas.nasa.gov/hecc/support/kb/60/